# Analysis of multi-server two-stage queueing network with split and blocking 

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#### Abstract

Parallel lines are often used to increase production rate in manufacturing system or model the multiple-product manufacturing systems. In parallel systems, the main line splits into several lines in parallel, and after some operations, they merge into a main line again. This paper provides an approximation technique for multi-server two-stage networks with splits and blocking which will be a building block for analysis of the parallel system with splits.


Keywords: queueing network, multi-product manufacturing system, multi-server, split configuration

## 1. INTRODUCTION

Parallel lines are often used to increase production rate in manufacturing system or model the multiple-product manufacturing systems. In parallel systems, the main line splits into several lines in parallel, and after some operations, they merge into a main line again. For example, three types of lens module installed in smart phone are assembled in their own assembly line. After finishing assembly, all lens modules are sent to the resolution inspection machines which can inspect all types of lens modules automatically. Defective parts are separated from the main line to be either repaired or scraped.
There is an extensive literature on the analysis of flow lines for modelling single-product manufacturing system. For a detailed list of references for approximations and their application to manufacturing system, see review papers Dallery and Gershwin (1992), Papadopoulos and Heavey (1996), Li et al. (2009) and monographs Buzacott and Shantikumar (1993), Gershwin (1994), Altiok (1997), Li and Meerkov (2009). The study of the systems with split and/or merge configuration for multiple-product manufacturing system is relatively scarce, e.g. see Helber (2000) for the system with splits, Helber and Jusic (2004) for the system with merge and Colledani et al. (2005), Li (2005), Li and Huang (2005), Liu and Li (2009) for the system with merge and splits. Most of the approximation methods for flow lines and the system with merge and/or splits are either based on decomposition or on aggregation. The principal procedure of decomposition methods is to decompose the long line into subsystems with two-station that are mathematically tractable, and derives a set of equations that determine the unknown parameters of each subsystem, and finally develops an algorithm to solve these equations. The basic idea of aggregation method is to replace a two-station system by a single equivalent station and reduce the number of stations in the network. Thus an analysis of queueing network with twostation and one buffer is a building block in analysis of long
line using decomposition or aggregation method. A few papers deal with a three-station merge system as a building block for an analysis of the system with merge configuration (see Tan (2001), Helber and Mehrtens (2003), Diamantidis et al. (2004), Diamantidis and Papadopoulos (2006)).

There is a tremendous literature on the analysis of flow lines with single server at each service stage. However, there is a little literature for analysis of the system with multiple servers (van Vuuren et al., 2005, Diamantidis et al., 2007). Recently, Shin and Moon (unpublished) presents a decomposition method for multi-server flow line with exponential service times and finite buffers. They use the subsystems with two service stations and two buffers to reflect the dependence between consecutive stages and improve the quality of approximation. To the best of our knowledge, there is little literature for analysis of the multiple-product manufacturing systems with multiple servers at each service station. As mentioned above, it is necessary to analyse two-stage networks for an approximation of such queueing networks using decomposition method. The purpose of this paper is to provide an approximation method for multi-server two-stage networks with split and blocking which will be a building block for analysis of multi-server-finite-buffer queueing networks with parallel lines and split configuration.

This paper is organized as follows. In Section 2, the model is described in detail. Approximation procedure is presented in Section 3. Numerical results and concluding remarks are presented in Sections 4 and 5, respectively.

## 2. MODEL DESCRIPTION

We consider the two-stage open queueing network in which the first stage is one queue with multiple servers and finite buffer and the second stage consists of $J$ multi-server queues with finite buffer. Two stages are linked in split configuration as depicted in Fig. 1. The queue (split queue) at the first stage is denoted by $S_{0}$ and is called node 0 and the queues at the
second stage will be denoted by $S_{i}$ called node $i, i=1,2, \ldots, J$ The node $i$ consists of a service station $M_{i}$ with $m_{i}$ parallel identical servers and a buffer $B_{i}$ of size $K_{i}$ and let $N_{i}=m_{i}+K_{i}$, $i=0,1,2, \ldots, J$. The index $i$ can be viewed as the sequential number of the nodes at the second stage runs from 1 to $J$ unless otherwise stated. The service time of each server at $M_{i}$ is exponentially distributed with rate $\mu_{i}, i=0,1,2, \ldots, J$. Customers arrive to node 0 from outside according to a Poisson process with rate $\lambda$. When an arrival finds no free position in the buffer $B_{0}$ (buffer is full), it is lost. After completion of the service at node 0 , the request proceeds to node $i$ with probability $p_{i}$ and is taken immediately into service if one of the servers at $M_{i}$ is free, otherwise is placed into the buffer $B_{i}$ and waits there until a server is available. If buffer $B_{i}$ is full at the instant of service completion at node 0 , the customer is forced to stay at node 0 until a space becomes available at node $i$. During this period, the server just completed its service is blocked. This type of blocking mechanism is called blocking-after-service (BAS). A customer who has completed its service at node $i$ leaves the system.


Fig. 1. A two-stage queueing network with split
Let $X_{0}$ be the number of customers at node 0 which includes the customers being served at $M_{0}$ and waiting requests at $B_{0}$ but does not include the customers blocked to enter the downstream node $i, i=1,2, \ldots, J$, in stationary state. By $X_{i}$ denote the number of customers at node $i$ that include the customers blocked at node 0 and $Y_{i}$ the number of customers who are blocked to enter into node $i, i=1,2, \ldots, J$. Note that the state space of $X_{i}$ is $\left\{0,1, \ldots, H_{i}\right\}$, where $H_{i}=N_{i}+m_{0}$.

## 3. APPROXIMATION

For an investigation of the behaviour of node $i$ and node 0 , we consider the two-node tandem queue denoted by $L_{i}$ where departures from the first stage to outside of the system are allowed as depicted in Fig.2. We denote the first stage and the second stage by $S_{0}$ and $S_{i}$, respectively. The stage $S_{0}$
consists of a service station $M_{0}$ with $m_{0}$ exponential servers with rate $\mu_{0}$ in parallel and a buffer $B_{0}$ of size $K_{0}$. The stage $S_{i}$ consists of a service station with $m_{i}$ exponential servers with rate $\mu_{i}$ in parallel and a finite buffer of size $K_{i}$. The customers at $S_{0}$ may be blocked. The BAS blocking rule is assumed in the system. We assume that each customer at $S_{0}$ is classified by type $j$ customer with probability $p_{i}, j=1,2, \ldots, J$ immediately after its service. Type $i$ customer joins $S_{i}$ and the customer of type $j, j \neq i$ leaves the system or is blocked after its service at
$S_{0}$. We assume that the blocking probability of type $j, j \neq i$ customer depends not only on the state $h$ of $S_{0}$ but also on the state $\boldsymbol{y}=\left(y_{1}, \ldots, y_{J}\right)$ of the number of blocked customers $\boldsymbol{Y}=\left(Y_{1}, \ldots, Y_{J}\right)$ and denote it by $\beta_{j}(h, \boldsymbol{y})$.


Fig. 2. Two-stage tandem queue $L_{i}$
Let $X_{0}^{*}(t)$ be the number of customers excluding the blocked one at $S_{0}$ and $X_{i}^{*}(t)$ the number of customers at $S_{i}$ and the customers blocked to enter $S_{i}$ at time $t$. Denote by $Y_{j}^{*}(t)$ the number of type $j$ customers blocked at $S_{0}$. Let

$$
\begin{aligned}
& \boldsymbol{Z}_{i}^{*}(t)=\left(X_{0}^{*}(t), X_{i}^{*}(t), \boldsymbol{Y}_{i}^{*}(t)\right), \\
& \boldsymbol{Y}_{i}^{*}(t)=\left(Y_{1}^{*}(t), \ldots, Y_{i-1}^{*}(t), Y_{i+1}^{*}(t), \ldots, Y_{J}^{*}(t)\right) .
\end{aligned}
$$

The stochastic process $\boldsymbol{Z}_{i}^{*}=\left\{\boldsymbol{Z}_{i}^{*}(t), t \geq 0\right\}$ is a Markov chain. By $X_{0}^{*}, X_{i}^{*}$ and $Y_{j}^{*}$ denote the stationary versions of $X_{0}^{*}(t), X_{i}^{*}(t)$ and $Y_{j}^{*}(t)$, respectively. Note that $0 \leq X_{0}^{*} \leq N_{0}$ and for a given $X_{0}^{*}=h$, we have that

$$
0 \leq X_{i}^{*} \leq N_{i}+m_{0}-\max \left(h-K_{0}, 0\right) .
$$

Note that $Y_{j}^{*}=\max \left(X_{i}^{*}-N_{i}, 0\right)$ and for a given $\left(X_{0}^{*}, X_{i}^{*}\right)$,

$$
0 \leq \boldsymbol{Y}_{i}^{*} \mathbf{e}=\sum_{k \neq i} Y_{k}^{*} \leq m_{0}-\max \left(X_{i}^{*}-N_{i}, 0\right)-\max \left(X_{0}^{*}-K_{0}, 0\right),
$$

where $\mathbf{e}=(1, \ldots, 1)^{\mathrm{T}}$ is column vector of appropriate size whose components are all 1. The state space of $\boldsymbol{Z}_{i}^{*}$ is $\boldsymbol{S}_{i}=\cup_{h=0}^{N_{0}} \boldsymbol{h}$, where $\boldsymbol{h}=\left\{(h, n, \boldsymbol{y}), 0 \leq n \leq N_{i}+m_{0}-\max \left(h-K_{0}, 0\right), 0 \leq \boldsymbol{y} \mathbf{e} \leq l(h, n)\right\}$ and $l(h, n)=m_{0}-\max \left(n-N_{i}, 0\right)-\max \left(h-K_{0}, 0\right)$.
The transition rates $q_{i}\left((h, n, \boldsymbol{y}),\left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)\right)$ of $\boldsymbol{Z}_{i}^{*}$ are as follows: for $h=0,1, \ldots, N_{0}-1$,

$$
q_{i}\left((h, n, \boldsymbol{y}),\left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)\right)= \begin{cases}\lambda, & \left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)=(h+1, n, \boldsymbol{y}), \\ \mu_{0}(h, n, \boldsymbol{y}) p_{i}, & \left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)=(h-1, n+1, \boldsymbol{y}), \\ \mu_{0}(h, n, \boldsymbol{y}) \sum_{j \neq i} p_{j} \bar{\beta}_{j}(h, \boldsymbol{y}), & \left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)=(h-1, n, \boldsymbol{y}) \\ \mu_{0}(h, n, \boldsymbol{y}) p_{j} \beta_{j}(h, \boldsymbol{y}), & \left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)=\left(h-1, n, \boldsymbol{y}+\mathbf{e}_{j}\right), j \neq i \\ \mu_{i}(n), & \left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)=(h, n-1, \boldsymbol{y}) \\ \mu_{j}\left(m_{j}\right), & \left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)=\left(h, n, \boldsymbol{y}-\mathbf{e}_{j}\right), j \neq i \\ 0, & \text { otherwise, }\end{cases}
$$

where $\mathbf{e}_{k}$ is the ( $J-1$ )-dimensional vector whose $k$ th component is 1 and others are all $0, \bar{\beta}_{j}(h, \boldsymbol{y})=1-\beta_{j}(h, \boldsymbol{y})$,

$$
\begin{aligned}
& \mu_{0}(h, n, \boldsymbol{y})=\min \left(h, m_{0}-\max \left(n-N_{i}, 0\right)-\boldsymbol{y e}\right) \mu_{0}, \\
& \mu_{j}(n)=\min \left(n, m_{j}\right) \mu_{j}, j=1,2, \ldots, J .
\end{aligned}
$$

Remark. One can consider the system whose routing probability $p_{j}$ may depend on the number $\boldsymbol{Y}=\boldsymbol{y}$ of customers blocked at node $j, j=1,2, \ldots, J$ by replacing $p_{j}$ with probability $p_{j}(\boldsymbol{y})$.

It can be easily seen that the Markov chain $\boldsymbol{Z}_{i}^{*}$ is a level dependent quasi-birth-and-death (LDQBD) process whose generator $Q_{i}$ is of the form

$$
\mathrm{Q}_{\mathrm{i}}=\left(\begin{array}{ccccc}
\mathrm{B}_{\mathrm{i}}^{(0)} \mathrm{A}_{\mathrm{i}}^{(0)} & & & \\
\mathrm{C}_{\mathrm{i}}^{(1)} \mathrm{B}_{\mathrm{i}}^{(1)} & \mathrm{A}_{\mathrm{i}}^{(1)} & & & \\
& \ddots & \ddots & \ddots & \\
& & \left.\mathrm{C}_{\mathrm{i}}^{\left(\mathrm{N}_{0}-1\right)} \mathrm{B}_{\mathrm{i}} \mathrm{~N}_{0}-1\right) & \mathrm{A}_{\mathrm{i}}^{\left(\mathrm{N}_{0}-1\right)} \\
& & & & \mathrm{C}_{\mathrm{i}}^{\left(\mathrm{N}_{0}\right)} \\
& \mathrm{B}_{\mathrm{i}}^{\left(\mathrm{N}_{0}\right)}
\end{array}\right),
$$

where the block matrix $B_{i}^{(h)}$ is the square matrix of size $|\boldsymbol{h}|$, the number of states of level $\boldsymbol{h}, h=0,1,2, \ldots, N_{0}$ and the components of matrices $A_{i}^{(h)}, B_{i}^{(h)}$ and $C_{i}^{(h)}$ are given in terms of $q_{i}\left((h, n, \boldsymbol{y}),\left(h^{\prime}, n^{\prime}, \boldsymbol{y}^{\prime}\right)\right)$. Let

$$
\begin{aligned}
& \pi_{i}(h, n, \boldsymbol{y})=\mathrm{P}\left(\left(X_{0}^{*}, X_{i}^{*}, \boldsymbol{Y}_{i}^{*}\right)=(h, n, \boldsymbol{y})\right),(h, n, \boldsymbol{y}) \in \boldsymbol{S}_{i} \\
& \boldsymbol{\pi}_{i}(h)=\left(\pi_{i}(h, n, \boldsymbol{y}),(h, n, \boldsymbol{y}) \in \boldsymbol{h}\right), h=0,1, . ., N_{0} .
\end{aligned}
$$

The stationary distribution $\boldsymbol{\pi}_{i}=\left(\pi_{i}(0), \pi_{i}(1), \ldots, \pi_{i}\left(N_{0}\right)\right)$ satisfies the following equations

$$
\pi_{i}(n)=\pi_{i}(n-1) R_{i}(n), n=1,2, \ldots, N_{0},
$$

where $\boldsymbol{R}_{i}(n)$ is obtained recursively as follows

$$
\begin{aligned}
& R_{i}\left(N_{0}\right)=A_{i}^{\left(N_{0}-1\right)}\left(-B_{i}^{\left(N_{0}\right)}\right)^{-1}, \\
& R_{i}(n)=A_{i}^{(n-1)}\left[-\left(B_{i}^{(n)}+R_{i}(n+1) C_{i}^{(n+1)}\right)\right]^{-1}, \\
& n=N_{0}-1, N_{0}-2, \ldots, 1 .
\end{aligned}
$$

The vector $\pi_{i}(0)$ is the unique solution of the equations

$$
\begin{aligned}
& \pi_{i}(0)\left(B_{i}^{(0)}+R_{i}(1) C_{i}^{(1)}\right)=0 \\
& \pi_{i}(0)\left[\mathbf{e}+\sum_{1 \leq n \leq N_{0}} R_{i}(1) R_{i}(2) \cdots R_{i}(n) \mathbf{e}\right]=1 .
\end{aligned}
$$

Let $\Psi$ be the number of busy servers at $M_{0}$. The mean arrival rate $\gamma_{i}(n)=\mu_{0} \mathrm{E}\left[\Psi \mid X_{i}=n\right]$ from node 0 to node $i$ given $X_{i}=n$. Once $\pi_{i}$ is given, we can calculate the approximation formulae for $\gamma_{i}(n)$ and $\beta_{i}(h, \boldsymbol{y})$ as follows:

$$
\begin{aligned}
\gamma_{i}(n)= & \mu_{0} p_{i} \mathrm{E}\left[\Psi \mid X_{i}^{*}=n\right] \\
= & \mu_{0} p_{i} \sum_{\left\{(h, \boldsymbol{y}):(h, n, \boldsymbol{y}) \in S_{i}^{*}(n)\right\}} E\left[\Psi_{i} \mid \boldsymbol{Z}_{i}^{*}=(n, h, \boldsymbol{y})\right] \\
& \times \operatorname{P}\left(X_{0}^{*}=h, \boldsymbol{Y}_{i}^{*}=\boldsymbol{y} \mid X_{i}^{*}=n\right) \\
= & \sum_{\left\{(h, \boldsymbol{y}):(h, n, \boldsymbol{y}) \in S_{i}^{*}(n)\right\}} \mu_{0}(h, n, \boldsymbol{y}) p_{i} \frac{\pi_{i}(h, n, \boldsymbol{y})}{P\left(X_{i}^{*}=n\right)},
\end{aligned}
$$

where $S_{i}^{*}(n)=\left\{(h, k, \boldsymbol{y}) \in S_{i}: k=n\right\}$,

$$
\begin{align*}
\beta_{i}(h, \boldsymbol{y}) & =\mathrm{P}\left(X_{i}^{*} \geq N_{i} \mid X_{0}^{*}=h, \boldsymbol{Y}=\boldsymbol{y}\right) \\
& = \begin{cases}\mathrm{P}\left(X_{i}^{*}=N_{i} \mid X_{0}^{*}=h, \boldsymbol{Y}=\boldsymbol{y}\right), & y_{i}=0 \\
1, & y_{i}>0\end{cases}  \tag{2}\\
& = \begin{cases}\pi_{i}\left(h, N_{i}, \boldsymbol{y}^{*}\right) / \sum_{n=0}^{N_{i}} \pi_{i}\left(h, n, \boldsymbol{y}^{*}\right), & y_{i}=0, \\
1, & y_{i}>0,\end{cases}
\end{align*}
$$

where $\boldsymbol{y}^{*}=\left(y_{1}, \ldots, y_{i-1}, y_{i+1}, \ldots, y_{J}\right)$.
Throughput from node $i$ is given by

$$
\begin{equation*}
T P_{i}=\sum_{(h, n, \boldsymbol{y}) \in S_{\boldsymbol{i}}} \mu_{i}(n) \pi_{i}(h, n, \boldsymbol{y}) \tag{3}
\end{equation*}
$$

We use the iterative algorithm to determine the parameters and calculate the throughput. Let $L_{i}^{(k)}, k=0,1,2, \ldots$ be the system $L_{i}$ in the $k t h$ iteration and by $\beta_{i}^{(k)}(h, y)$ and $\gamma_{i}^{(k)}(n)$ denote $\beta_{i}(h, \boldsymbol{y})$ and $\gamma_{i}(n)$ corresponding to the subsystem $L_{i}^{(k)}$.

## Algorithm

## Step 0 [Initial Step]

Stage 1. Initially we assume that no type $j$ customers $(j=2, \ldots, J)$ at node $S_{0}$ are blocked and consider the system $L_{1}^{(0)}$ as two node tandem queue where the type $j$ customers $(j=2, \ldots, J)$ leave the system immediately after completing their service. Calculate $\pi_{1}{ }^{(0)}(h, n, \mathbf{0})$ with $\beta_{j}^{(0)}(h, \mathbf{0})=0, \quad j=2, \ldots, J$ and then calculate $\left.\beta_{1}{ }^{(0)}\left(h, y_{1}, 0, \ldots, 0\right)\right)$ and $\gamma_{1}^{(0)}(n)$ using (1) and (2), respectively.

Stage $\boldsymbol{i}$. For $i=2,3, \ldots, J$, calculate $\pi_{i}^{(0)}\left(h, n, \boldsymbol{y}_{i}{ }^{*}\right)$ with $\beta_{j}^{(0)}(h$, $\left.y_{1}, \ldots, y_{j}, 0, \ldots, 0\right), j=1,2, \ldots, i-1$ and $\beta_{j}^{(0)}(h, \mathbf{0})=0, j=i+1, \ldots, J$ and then calculate $\beta_{i}^{(0)}\left(h, y_{1}, \ldots, y_{i}, 0, \ldots, 0\right)$ and $\gamma_{i}^{(0)}(n)$ using (1) and (2), respectively.
Step 1 For $i=1,2, \ldots, J$, calculate the stationary distribution $\pi_{i}^{(k)}$ of $L_{i}^{(k)}$ with $\beta_{j}^{(k)}(h, y), j<i$ and $\quad \beta_{j}^{(k-1)}(h, y), j>i$. Update $\beta_{i}^{(k)}(h, y)$ and $\gamma_{i}^{(k)}(n)$ using (1) and (2), respectively.

Step 2 [Stopping Criterion] Check the stopping criterion

$$
T O L=\max _{i, h, \boldsymbol{y}}\left|\gamma_{i}^{(k)}(n)-\gamma_{i}^{(k-1)}(n)\right|<\varepsilon,
$$

where $\varepsilon>0$ is a given tolerance. If the stopping condition is satisfied, then stop the iteration and calculate the throughput from node $i$ using (3). Otherwise, go to Step 1.

No analytical proof of convergence or accuracy can be given for this algorithm. However extensive numerical experiments show the convergence of the algorithm. The complexity of this method is as follows. Within the iterative algorithm, calculating $\pi_{i}(n)$ consumes most of the time. For solving the subsystem $L_{i}$, inversion of $N_{0}+1$ matrices of size $|\boldsymbol{h}|, 0 \leq h \leq N_{0}$ is required. The number of iterations needed is difficult to predict because it depends on the tolerance $\varepsilon$ and the length of the line and system parameters.

## 4. NUMERICAL RESULTS

In this section, some numerical results are presented for the performance of approximations. We consider the system with parameters $m_{0}=3, m_{1}=1, m_{2}=2, m_{3}=3$ and the service rates $\mu_{0}=1.0, \mu_{i}=1.0 / m_{i}, i=1,2,3$. The approximation results (App) for throughput, mean $\mathrm{E}\left[X_{i}\right]$ and standard deviation $\operatorname{Std}\left[X_{i}\right]$ of the number of customers at node $i$ are numerically compared with those of simulation (Sim) in Tables 1-3 for $K_{0}=5$ and arrival rates $\lambda=1.0$ and $\lambda=3.0$. Table 1 represents the results for the system with $K_{i}=0, i=1,2,3$. Similarly, Tables 2-3 present the results for the system with $K_{i}=2, i=1,2,3$ and
$K_{i}=4-i, \quad i=1,2,3$, respectively. The maximal number of iterations of the algorithm to obtain each result for the tolerance $\varepsilon=10^{-5}$ in the examples is 5 for $\lambda=3.0$.

Simulation models are developed with ARENA (Kelton et al., 1998). Simulation run time is set to 11,000 unit times including 1,000 unit times of warm-up period. Each simulation is repeated sufficiently many times such that the half length of $95 \%$ confidence interval for throughput and $\mathrm{E}\left[X_{i}\right]$ are smaller than $1 \%$. The relative percentage error is calculated by $\operatorname{Err}(\%)=($ App-Sim $) \times 100 /$ Sim. Numerical results show the approximation works very well.

Table 1. Comparisons with Simulation I ( $K_{i}=0, i=1,2,3$ )

| $\lambda$ | node | Throughput |  |  | $\mathrm{E}\left[X_{i}\right]$ |  |  | $\operatorname{Std}\left[X_{i}\right]$ |  |  |
| ---: | :---: | ---: | ---: | ---: | ---: | :---: | ---: | ---: | ---: | ---: |
|  |  | App | Sim | $\operatorname{Err}(\%)$ | App | Sim | $\operatorname{Err}(\%)$ | App | Sim | $\operatorname{Err}(\%)$ |
| 1.0 | 0 | 0.995 | 0.996 | 0.1 | 1.175 | 1.182 | 0.6 | 1.237 | 1.244 | -0.6 |
|  | 1 | 0.199 | 0.200 | 0.7 | 0.246 | 0.247 | 0.4 | 0.545 | 0.547 | -0.3 |
|  | 2 | 0.298 | 0.297 | -0.4 | 0.650 | 0.648 | -0.2 | 0.861 | 0.860 | 0.2 |
|  | 3 | 0.497 | 0.498 | 0.2 | 1.661 | 1.675 | 0.8 | 1.389 | 1.396 | -0.5 |
| 3.0 | 0 | 1.681 | 1.675 | -0.4 | 5.488 | 5.489 | 0.0 | 1.500 | 1.494 | 0.4 |
|  | 1 | 0.336 | 0.336 | -0.1 | 0.489 | 0.495 | 1.1 | 0.795 | 0.802 | -0.8 |
|  | 2 | 0.504 | 0.502 | -0.5 | 1.257 | 1.244 | -1.0 | 1.241 | 1.240 | 0.1 |
|  | 3 | 0.841 | 0.837 | -0.4 | 3.410 | 3.409 | 0.0 | 1.598 | 1.605 | -0.4 |

Table 2. Comparisons with Simulation II ( $K_{i}=2, i=1,2,3$ )

| $\lambda$ | node | Throughput |  |  | $\mathrm{E}\left[X_{i}\right]$ |  |  | $\operatorname{Std}\left[X_{i}\right]$ |  |  |
| :---: | :---: | :---: | :---: | ---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | App | Sim | $\operatorname{Err}(\%)$ | App | Sim | $\operatorname{Err}(\%)$ | App | Sim | $\operatorname{Err}(\%)$ |
| 1.0 | 0 | 0.999 | 1.000 | 0.1 | 1.069 | 1.078 | 0.9 | 1.114 | 1.124 | -0.9 |
|  | 1 | 0.200 | 0.201 | 0.4 | 0.250 | 0.249 | -0.2 | 0.558 | 0.557 | 0.1 |
|  | 2 | 0.300 | 0.299 | -0.3 | 0.658 | 0.660 | 0.4 | 0.885 | 0.891 | -0.7 |
|  | 3 | 0.499 | 0.501 | 0.2 | 1.715 | 1.724 | 0.5 | 1.512 | 1.511 | 0.1 |
| 3.0 | 0 | 1.872 | 1.868 | -0.2 | 5.303 | 5.304 | 0.0 | 1.731 | 1.732 | 0.0 |
|  | 1 | 0.374 | 0.376 | 0.3 | 0.625 | 0.634 | 1.3 | 1.004 | 1.017 | -1.3 |
|  | 2 | 0.562 | 0.560 | -0.4 | 1.632 | 1.653 | 1.3 | 1.631 | 1.664 | -1.9 |
|  | 3 | 0.936 | 0.933 | -0.4 | 5.156 | 5.109 | -0.9 | 1.991 | 2.025 | -1.7 |

Table 3. Comparisons with Simulation III ( $K_{1}=3, K_{2}=2, K_{3}=1$ )

| $\lambda$ | node | Throughput |  |  | $\mathrm{E}\left[X_{i}\right]$ |  |  | $\operatorname{Std}\left[X_{i}\right]$ |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | App | Sim | $\operatorname{Err}(\%)$ | App | $\operatorname{Sim}$ | $\operatorname{Err}(\%)$ | App | Sim | $\operatorname{Err}(\%)$ |
| 1.0 | 0 | 0.998 | 0.998 | 0.0 | 1.092 | 1.089 | 0.2 | 1.142 | 1.136 | 0.6 |
|  | 1 | 0.200 | 0.200 | -0.1 | 0.250 | 0.250 | -0.2 | 0.559 | 0.557 | 0.2 |
|  | 2 | 0.299 | 0.299 | 0.0 | 0.657 | 0.663 | -0.9 | 0.885 | 0.891 | -0.7 |
|  | 3 | 0.499 | 0.499 | 0.0 | 1.697 | 1.688 | 0.5 | 1.465 | 1.451 | 1.0 |
| 3.0 | 0 | 1.837 | 1.824 | 0.7 | 5.330 | 5.340 | -0.2 | 1.697 | 1.688 | 0.5 |
|  | 1 | 0.367 | 0.366 | 0.3 | 0.619 | 0.627 | -1.4 | 1.023 | 1.045 | -2.1 |
|  | 2 | 0.551 | 0.544 | 1.2 | 1.586 | 1.585 | 0.1 | 1.610 | 1.626 | -1.0 |
|  | 3 | 0.918 | 0.913 | 0.6 | 4.440 | 4.433 | 0.2 | 1.749 | 1.766 | -1.0 |

## 5. CONCLUSION

In this paper, we developed an approximation method for a multi-server two-stage queueing network with finite buffers and split. The system is decomposed into subsystems that have two service stations with multiple servers and two buffers and departures from the first stage to outside of the system are allowed. An iterative algorithm is presented for calculating the parameters. The approximation method proposed can be used to analyse the complex queueing
networks with split configuration. Furthermore, the analysis of multi-server system can be used for allocation of the servers and buffers in design of manufacturing systems.

The analysis of the system with more complex splitting rule, the server failure, more complex arrival process and/or more general service distribution than exponential and the long system with split configuration are left as future research area.

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